

MATLAB과 함께하는 딥러닝 4주 완성 부트캠프

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세션3. 딥러닝 프로젝트를 위한 데이터 준비 기법 MATLAB과 함께하는 딥러닝 4주 완성 부트캠프

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Accelerating the pace of engineering and science

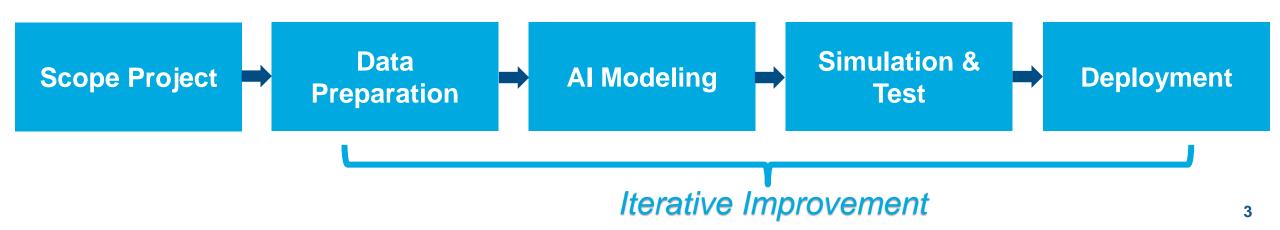


Software System Design

Traditional Software

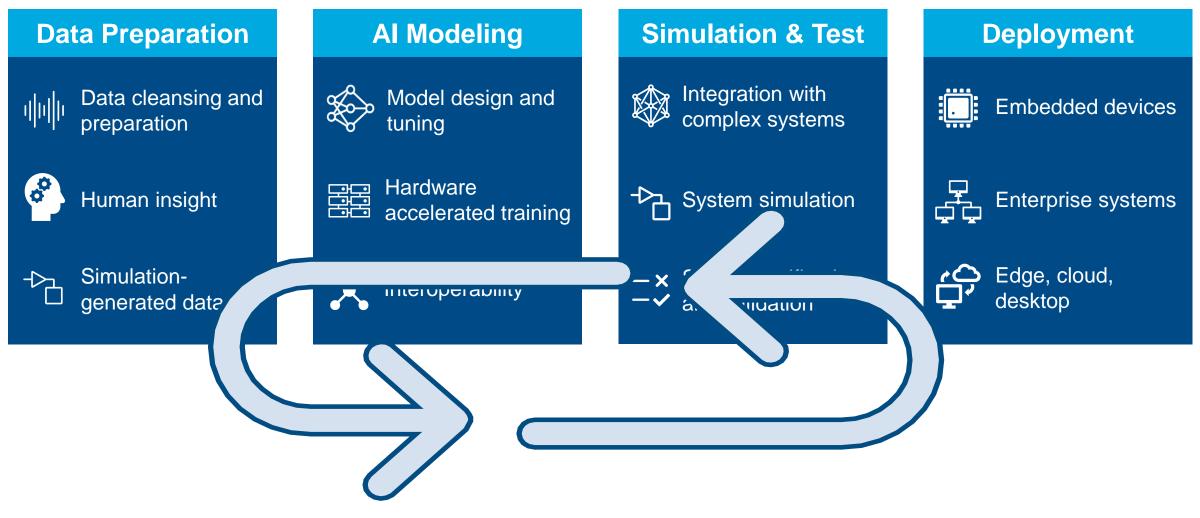


 Al Software = Code + Data (Model/Algorithm)





Al-driven system design

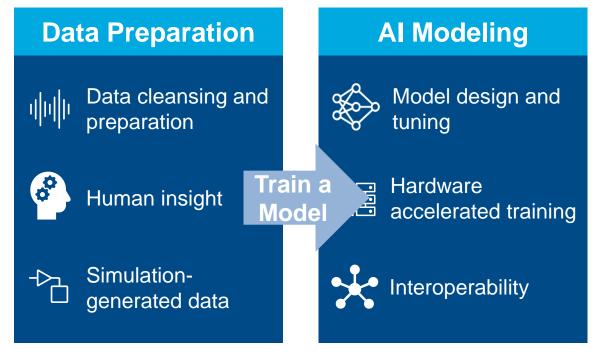


Iterative refinement for system improvement



Making it systematic : MLOps





- Error Analysis
 - Model-centric
 - How can I tune the model architecture?
 - Data-centric
 - How can I modify my data?
- Data-centric AI
 - Error analysis to identify the types of data the algorithm does poorly on
 - Hold the code fixed and iteratively improve the data.



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Data preparation represents most of your AI effort...

Transforming raw data for useful modeling and analysis is a critical step.

Example : Clean vs. Noisy data



Source: Andrew Ng slide from MLOps:From Model-centric to Data-centric AI 2021

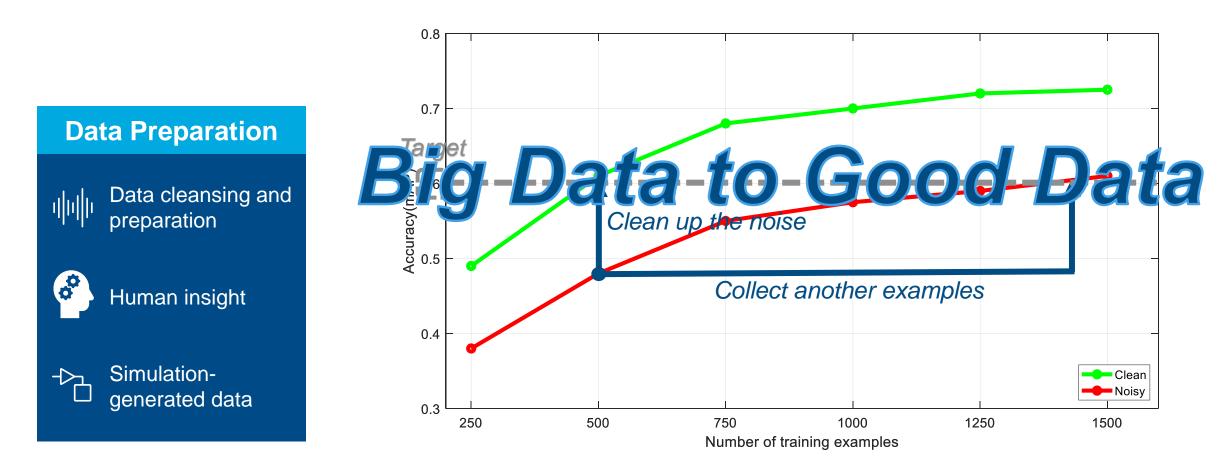


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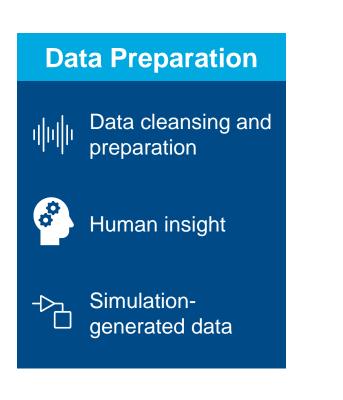


From Big data to Good Data

Transforming raw data for useful modeling and analysis is a critical step.

Data

(inputs x, labels y) Big Data to Good Data

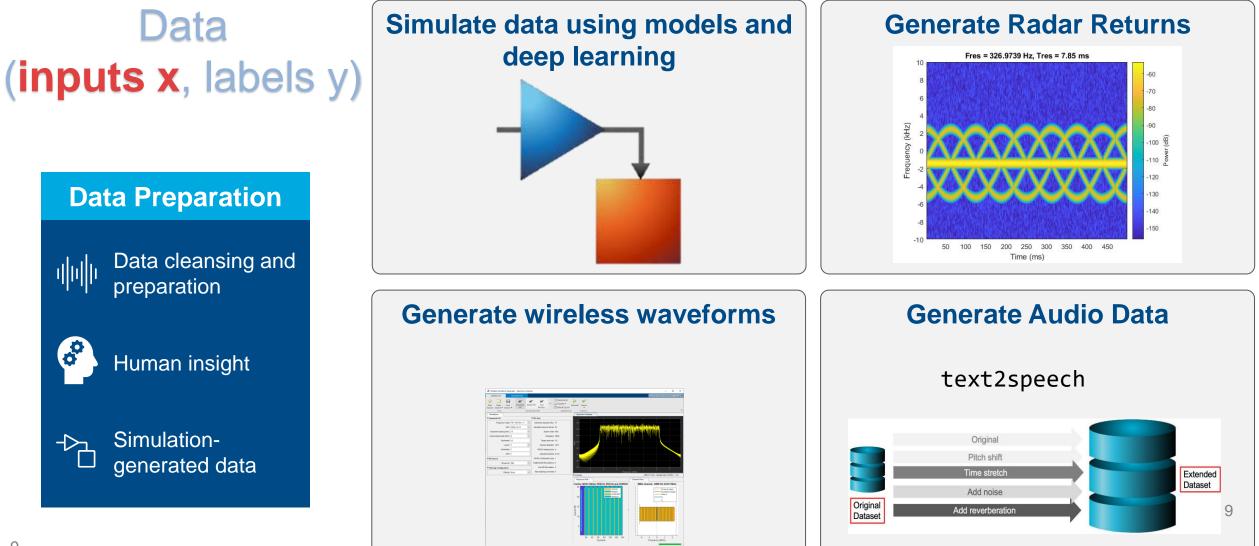


- Good data is:
 - Defined consistently
 - Definition of labels y is unambiguous
 - Cover of important cases
 - Good coverage of inputs x
 - Has timely feedback from production data
 - Distribution covers data drift and concept drift
 - Sized appropriately

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Synthetic Data Generation and Augmentation to deal with less data

Simulate your system for synthetic data generation





Synthetic Data Generation and Augmentation to deal with less data

Simulate your system for synthetic data generation

Data (inputs x, labels y)

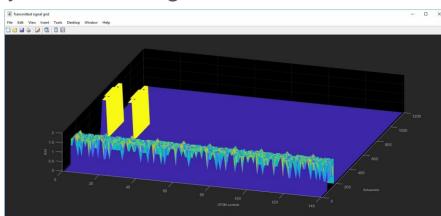
Data Preparation

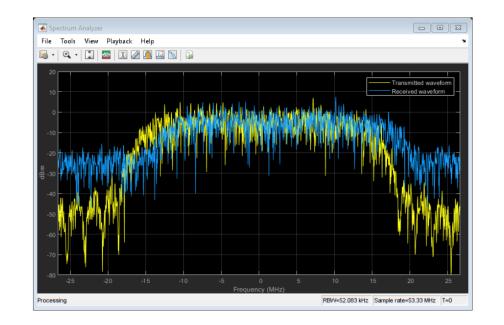
IIIII Data cleansing and preparation



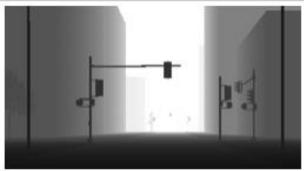
Human insight

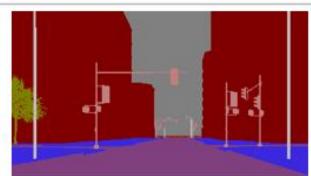
->> Simulationgenerated data







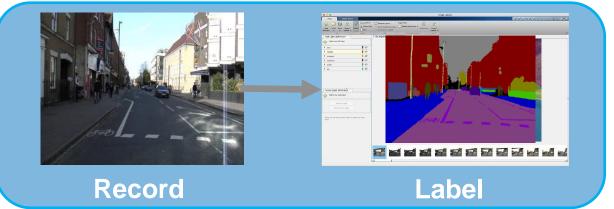




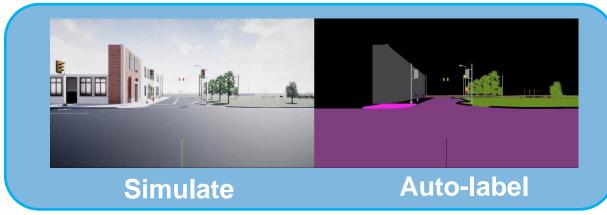


One example of leveraging simulation for data synthesis

Traditional deep learning workflow



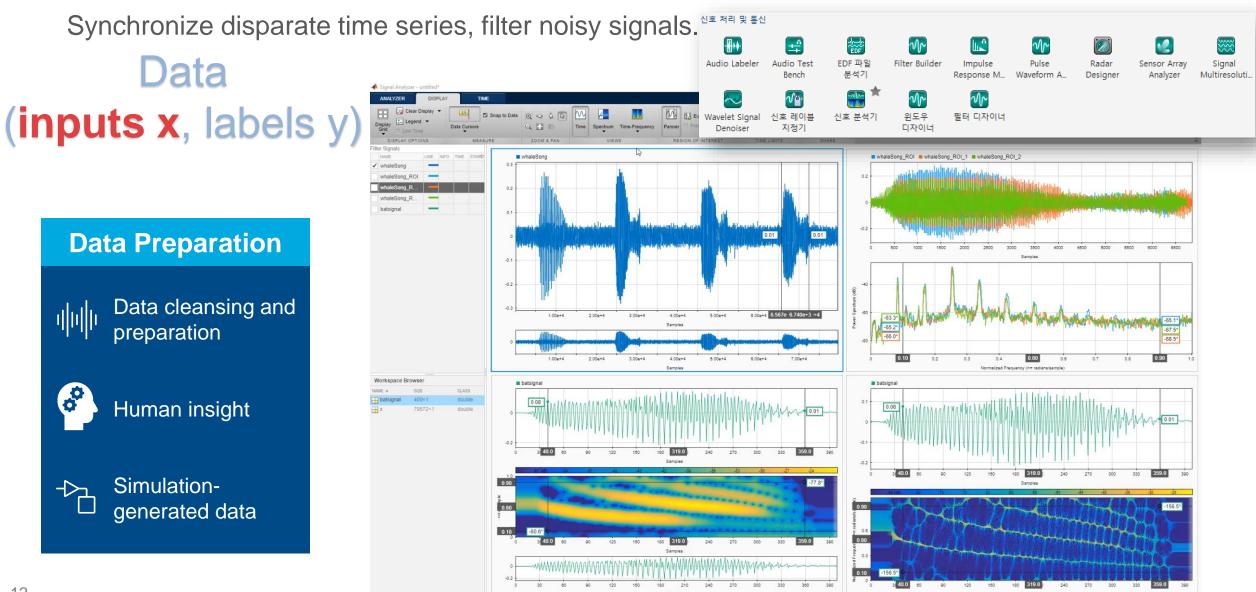
Simulation-based workflow





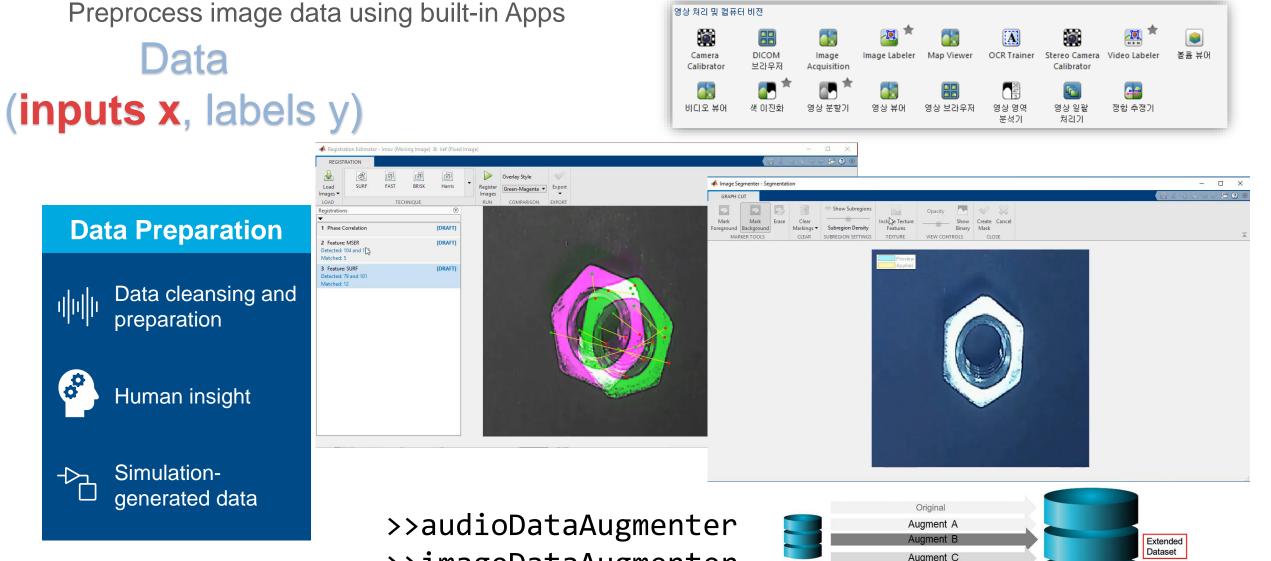
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Spend less time preprocessing and labeling data





Spend less time preprocessing and labeling data



영상 처리 및 컴퓨터 비전

Original Dataset

>>imageDataAugmenter



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Spend less time preprocessing and labeling data

Automate labeling of Lidar, image, video, and signal.

Data (inputs x, **labels y**)



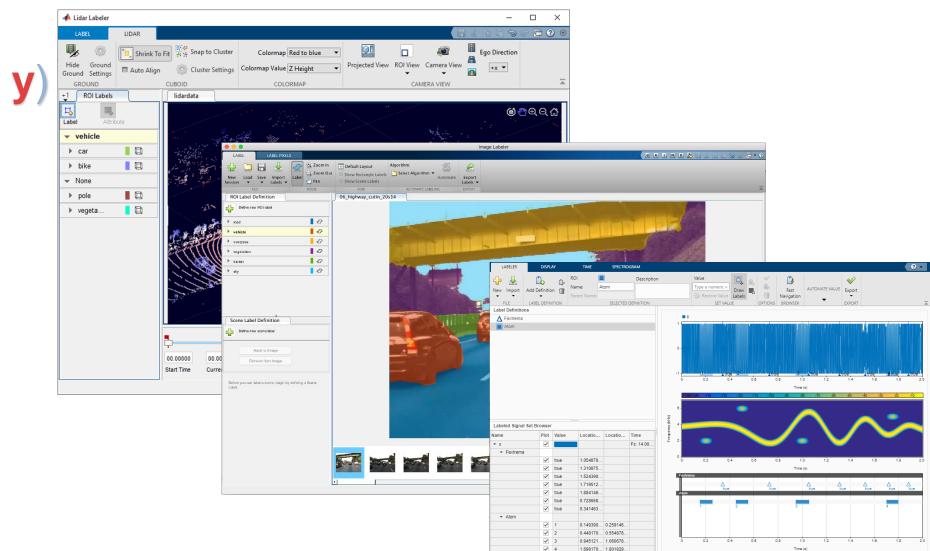
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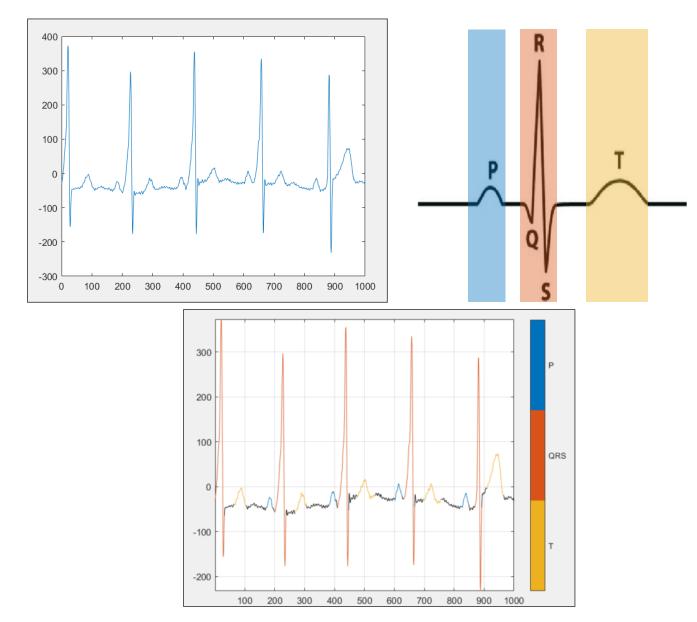
Human insight

Simulationgenerated data





Example1 : Labeling key regions of ECG waveforms



- ECG signals contain P,QRS and T waves.
- Identifying these can help with diagnosis and classification.
- Dataset contains 210 ECG signals ,~ 15 minutes long, labeled by cardiologist



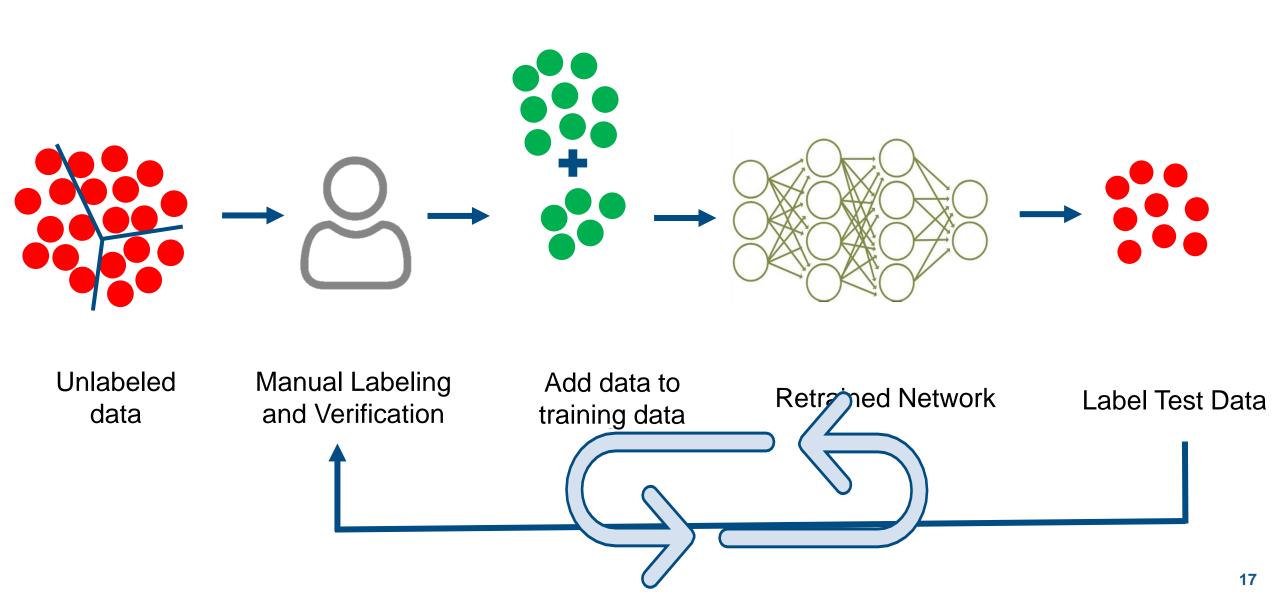
Example2 : Labeling vehicles for object detection



- Video data from my black box
- Wants to label cars in every frames of the video
- Do not have experiences on custom algorithm to detect vehicle
- Attended MATLAB Deep Learning Bootcamp #1



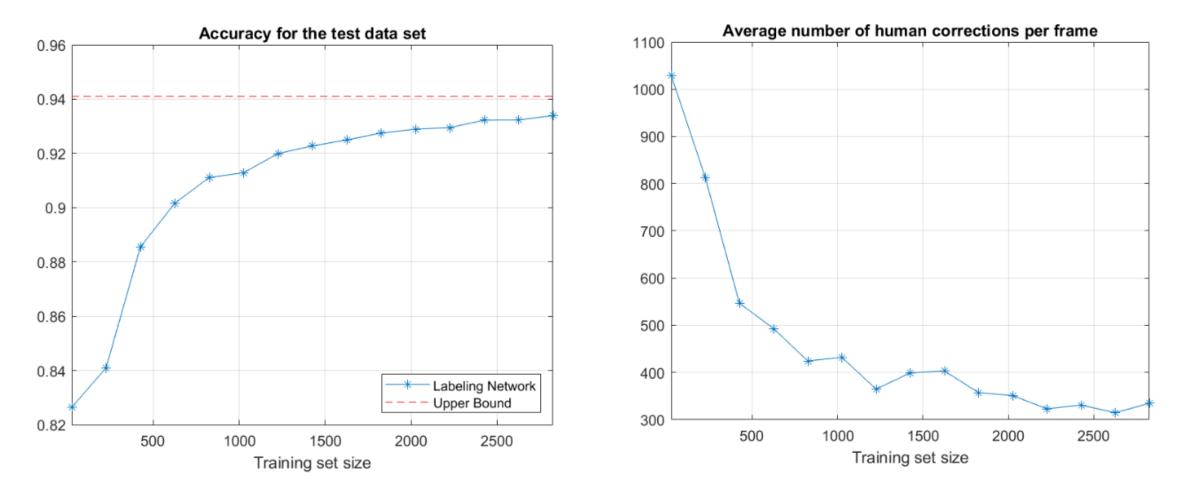
Establish iterative labeling process with deep learning





Result of iterative learning process

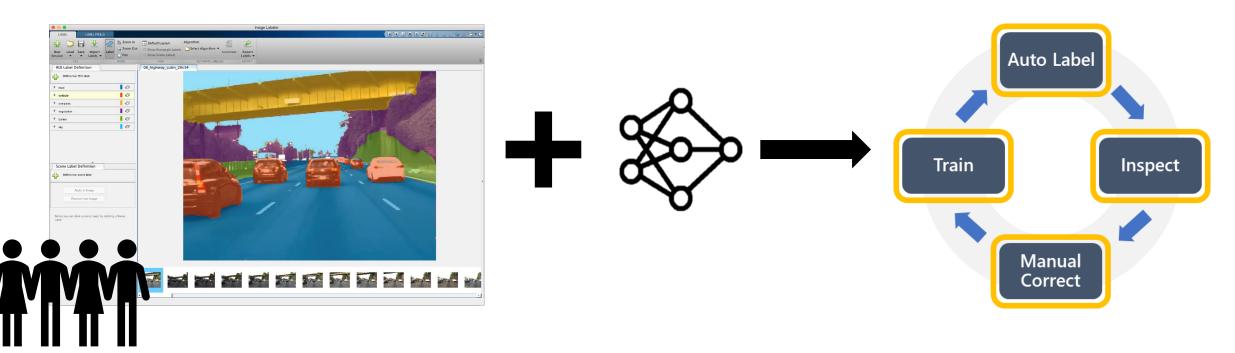
Iteratively the accuracy increases, needs for human correction decreases.





Strategy for automated labeling and iterative learning

- Manually label with colleagues
- Use pretrained model for labeling automation
- Establish iterative labeling process





HYUNDAI STEEL uses pixel-based deep learning technique for steel material image

HYUNDAI STEEL – HYUNDAI MOTOR GROUP



HYUNDAI STEEL developed pixel-based steel material image analysis application using deep learning and established labeling automation, enhancement technique with MATLAB.

In the field, analysis engineers using dot method, which manually analyze images from an optical microscope but it takes lots of time and strongly depends on the experience of the individual analyst.

With MATLAB, they created labeling automation algorithm using unsupervised machine learning technique and integrated it to labeling app for automation. By iterating labeling data based on machine learning and training network with deep learning model, they've enhanced prediction accuracy up to 85% which can fairly acceptable for replacing current analysis workflow.

Advantages of using MATLAB and Simulink:

- Leveraged high-level APIs for applying image preprocessing techniques
- Interactive Apps enabled non-AI expert can do deep learning project
- Golden reference examples for quick prototyping
- Leverage MathWorks support to maximize the benefits of using MATLAB

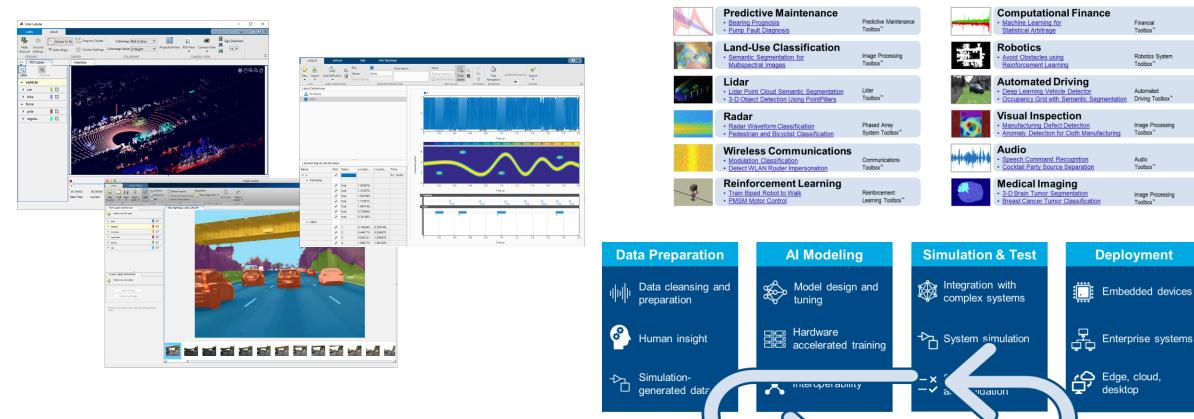
Even though I had limited knowledge on Image processing and deep learning, I could successfully adopt deep learning for my project. With evaluation support from MathWorks, we could prototype our approach easily with limited time bound.



Semantic Segmentation Overview (\mathbf{b}) Watch video



Build Systematic Data Management Process



- App based labeling tools
- Application specific preprocessing functions
- Deployment workflow for timely feedback from production data

MathWorks: helping engineers & scientists build Deep Learning solutions



The Platform

MATLAB, Simulink, and over 100 add-on products for specialized applications



Your People

Helping you build an agile workforce today and preparing tomorrow's engineers



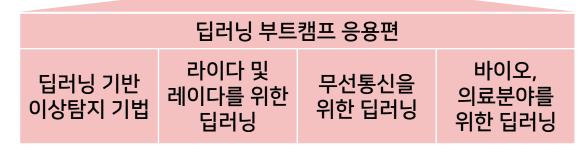
Our Expertise

From onboarding and implementation to solving advanced engineering challenges



<u>https://bit.ly/3hfSm24</u> 오늘 등록하세요! MathWorks







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